

Interactive theorem proving,  
automated reasoning,  
and dynamical systems

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# Formal methods

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“Formal methods” = logic-based methods in CS, in:

- automated reasoning
- hardware and software verification
- artificial intelligence
- databases

# Formal methods

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Based on logic and formal languages:

- syntax: terms, formulas, connectives, quantifiers, proofs
- semantics: truth, validity, satisfiability, reference

They can be used for:

- finding things (SAT solvers, constraint solvers, database query languages)
- proving things (automated theorem proving, model checking)
- verifying correctness (interactive theorem proving)

## Formal verification in industry

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- Intel and AMD use ITP to verify processors.
- Microsoft uses formal tools such as Boogie and SLAM to verify programs and drivers.
- Xavier Leroy has verified the correctness of a C compiler.
- Airbus uses formal methods to verify avionics software.
- Toyota uses formal methods for hybrid systems to verify control systems.
- Formal methods were used to verify Paris' driverless line 14 of the Metro.
- The NSA uses (it seems) formal methods to verify cryptographic algorithms.

# Formal verification in mathematics

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There is no sharp line between industrial and mathematical verification:

- Designs and specifications are expressed in mathematical terms.
- Claims rely on background mathematical knowledge.

Mathematics is more interesting:

- Problems are conceptually deeper, less heterogeneous.
- More user interaction is needed.

# Formal methods in mathematics

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“Conventional” computer-assisted proof:

- carrying out long, difficult, computations
- proof by exhaustion

Formal methods for discovery:

- finding mathematical objects
- finding proofs

Formal methods for verification:

- verifying ordinary mathematical proofs
- verifying computations.

# Formal methods in mathematics

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## Questions:

- How can computers help us reason about dynamical systems?
- How can computers help make results and computations more reliable?

# Outline

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- Formal methods
- Interactive theorem proving
- Automated reasoning
- Verified computation

## Interactive theorem proving

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Working with a proof assistant, users construct a formal axiomatic proof.

In most systems, this proof object can be extracted and verified independently.

## Interactive theorem proving

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Some systems with large mathematical libraries:

- Mizar (set theory)
- HOL (simple type theory)
- Isabelle (simple type theory)
- HOL light (simple type theory)
- Coq (constructive dependent type theory)
- ACL2 (primitive recursive arithmetic)
- PVS (classical dependent type theory)

## Interactive theorem proving

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Some theorems formalized to date:

- the prime number theorem
- the four-color theorem
- the Jordan curve theorem
- Gödel's first and second incompleteness theorems
- Dirichlet's theorem on primes in an arithmetic progression
- Cartan fixed-point theorems

There are good libraries for elementary number theory, real and complex analysis, point-set topology, measure-theoretic probability, abstract algebra, Galois theory, ...

## Interactive theorem proving

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Georges Gonthier and coworkers verified the Feit-Thompson Odd Order Theorem in Coq.

- The original 1963 journal publication ran 255 pages.
- The formal proof is constructive.
- The development includes libraries for finite group theory, linear algebra, and representation theory.

The project was completed on September 20, 2012, with roughly

- 150,000 lines of code,
- 4,000 definitions, and
- 13,000 lemmas and theorems.

## Interactive theorem proving

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Hales announced the completion of the formal verification of the Kepler conjecture (*Flyspeck*) in August 2014.

- Most of the proof was verified in HOL light.
- The classification of tame graphs was verified in Isabelle.
- Verifying several hundred nonlinear inequalities required roughly 5000 processor hours on the Microsoft Azure cloud.

## Interactive theorem proving

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Fabian Immler is working on verifying properties of dynamical systems in Isabelle.

- Proved existence and uniqueness of solutions to ODE's (Picard-Lindelöf and variations).
- With code extraction, can compute solutions. Following Tucker, has verified enclosures for the Lorenz attractor.

## Interactive theorem proving

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Johannes Hölzl, Luke Serafin, and I have verified the central limit theorem in Isabelle.

The proof relied on Isabelle's libraries for analysis, topology, measure theory, measure-theoretic probability.

We proved:

- the portmanteau theorem (characterizations of weak convergence)
- Skorohod's theorem
- properties of characteristic functions and convolutions
- properties of the normal distribution
- the Levy uniqueness theorem
- the Levy continuity theorem

# Interactive theorem proving

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**theorem** (in *prob\_space*) *central\_limit\_theorem*:

**fixes**

$X :: \text{"nat} \Rightarrow \text{'a} \Rightarrow \text{real}"$  **and**

$\mu :: \text{"real measure"}$  **and**

$\sigma :: \text{real}$  **and**

$S :: \text{"nat} \Rightarrow \text{'a} \Rightarrow \text{real}"$

**assumes**

$X\_indep: \text{"indep\_vars } (\lambda i. \text{borel}) X \text{ UNIV}"$  **and**

$X\_integrable: \text{"}\bigwedge n. \text{integrable } M (X n)"$  **and**

$X\_mean\_0: \text{"}\bigwedge n. \text{expectation } (X n) = 0"$  **and**

$\sigma\_pos: \text{"}\sigma > 0"$  **and**

$X\_square\_integrable: \text{"}\bigwedge n. \text{integrable } M (\lambda x. (X n x)^2)"$  **and**

$X\_variance: \text{"}\bigwedge n. \text{variance } (X n) = \sigma^2"$  **and**

$X\_distrib: \text{"}\bigwedge n. \text{distr } M \text{ borel } (X n) = \mu"$

**defines**

$S n \equiv \lambda x. \sum_{i < n}. X i x$

**shows**

$\text{"weak\_conv\_m } (\lambda n. \text{distr } M \text{ borel } (\lambda x. S n x / \text{sqrt } (n * \sigma^2)))$   
 $\text{(density lborel std\_normal\_density)"}$

# The Lean Theorem Prover

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Lean is a new interactive theorem prover, developed principally by Leonardo de Moura at Microsoft Research, Redmond.

It was “announced” in the summer of 2015.

It is open source, released under a permissive license, Apache 2.0.

See <http://leanprover.github.io>.

# The Lean Theorem Prover

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The aim is to bring interactive and automated reasoning together, and build

- an interactive theorem prover with powerful automation
- an automated reasoning tool that
  - produces (detailed) proofs,
  - has a rich language,
  - can be used interactively, and
  - is built on a verified mathematical library.

# The Lean Theorem Prover

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## Goals:

- Verify hardware, software, and hybrid systems.
- Verify mathematics.
- Combine powerful automation with user interaction.
- Support reasoning and exploration.
- Support formal methods in education.
- Create an eminently powerful, usable system.
- Bring formal methods to the masses.

# The Lean Theorem Prover

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Notable features:

- based on a powerful dependent type theory
- written in C++, with multi-core support
- small, trusted kernel with an independent type checker
- standard and HoTT instantiations
- powerful elaborator
- can use proof terms or tactics
- Emacs mode with proof-checking on the fly
- browser version runs in javascript
- already has a respectable library
- automation is now the main focus

# The Lean Theorem Prover

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```
structure semigroup [class] (A : Type) extends has_mul A :=
  (mul_assoc :  $\forall a b c, \text{mul} (\text{mul } a b) c = \text{mul } a (\text{mul } b c)$ )
```

```
structure monoid [class] (A : Type)
  extends semigroup A, has_one A :=
  (one_mul :  $\forall a, \text{mul one } a = a$ ) (mul_one :  $\forall a, \text{mul } a \text{ one} = a$ )
```

```
definition pow {A : Type} [s : monoid A] (a : A) :  $\mathbb{N} \rightarrow A$ 
| 0      := 1
| (n+1) := pow n * a
```

```
theorem pow_add (a : A) (m :  $\mathbb{N}$ ) :  $\forall n, a^{(m + n)} = a^m * a^n$ 
| 0      := by rewrite [nat.add_zero, pow_zero, mul_one]
| (succ n) := by rewrite [add_succ, *pow_succ, pow_add,
                          mul.assoc]
```

```
definition int.linear_ordered_comm_ring [instance] :
  linear_ordered_comm_ring int := ...
```

# The Lean Theorem Prover

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```
theorem sqrt_two_irrational {a b : ℕ} (co : coprime a b) :
  a^2 ≠ 2 * b^2 :=
  assume H : a^2 = 2 * b^2,
  have even (a^2),
    from even_of_exists (exists.intro _ H),
  have even a,
    from even_of_even_pow this,
  obtain (c : ℕ) (aeq : a = 2 * c),
    from exists_of_even this,
  have 2 * (2 * c^2) = 2 * b^2,
    by rewrite [-H, aeq, *pow_two, mul.assoc, mul.left_comm c],
  have 2 * c^2 = b^2,
    from eq_of_mul_eq_mul_left dec_trivial this,
  have even (b^2),
    from even_of_exists (exists.intro _ (eq.symm this)),
  have even b,
    from even_of_even_pow this,
  have 2 | gcd a b,
    from dvd_gcd (dvd_of_even 'even a') (dvd_of_even 'even b'),
  have 2 | 1,
    by rewrite [gcd_eq_one_of_coprime co at this]; exact this,
  show false,
    from absurd '2 | 1' dec_trivial
```

# The Lean Theorem Prover

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```
theorem is_conn_susp [instance] (n :  $\mathbb{N}_{-2}$ ) (A : Type)
  [H : is_conn n A] : is_conn (n .+1) (susp A) :=
is_contr.mk (tr north)
begin
  apply trunc.rec, fapply susp.rec,
  { reflexivity },
  { exact (trunc.rec ( $\lambda a$ , ap tr (merid a)) (center (trunc n A))) },
  { intro a, generalize (center (trunc n A)),
    apply trunc.rec, intro a', apply pathover_of_tr_eq,
    rewrite [transport_eq_Fr, idp_con],
    revert H, induction n with [n, IH],
    { intro H, apply is_prop.elim },
    { intros H,
      change ap (@tr n .+2 (susp A)) (merid a) = ap tr (merid a'),
      generalize a',
      apply is_conn_fun.elim n
        (is_conn_fun_from_unit n A a)
        ( $\lambda x$  : A, trunc.type.mk' n
          (ap (@tr n .+2 (susp A)) (merid a) = ap tr (merid x))),
      intros,
      change ap (@tr n .+2 (susp A)) (merid a) = ap tr (merid a),
      reflexivity } }
end
```

# Outline

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- Formal methods
- Interactive theorem proving
- Automated reasoning
- Verified computation

# Automated reasoning

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Ideal: given an assertion,  $\varphi$ , either

- provide a proof that  $\varphi$  is true (or valid), or
- give a counterexample

Dually: given some constraints either

- provide a solution, or
- prove that there aren't any.

In the face of undecidability:

- search for proofs
- search for solutions

# Automated reasoning

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Some fundamental distinctions:

- Domain-general methods vs. domain-specific methods
- Decision procedures vs. search procedures
- “Principled” methods vs. heuristics

# Automated reasoning

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## Domain-general methods:

- Propositional theorem proving
- First-order theorem proving
- Equational reasoning
- Higher-order theorem proving
- Nelson-Oppen “combination” methods

## Domain-specific methods:

- Linear arithmetic (integer, real, or mixed)
- Nonlinear real arithmetic (real closed fields, transcendental functions)
- Algebraic methods (such as Gröbner bases)

# Automated reasoning

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Formal methods in analysis:

- domain specific: reals, integers
- can emphasize either performance or rigor
- can get further in restricted domains

Methods:

- quantifier elimination for real closed fields
- linear programming, semidefinite programming
- combination methods
- numerical methods
- heuristic symbolic methods

## Symbolic methods

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Decision procedures for real closed fields represent a symbolic approach.

Problems:

- Complexity overwhelms.
- Polynomials may not be expressive enough.
- Undecidability sets in quickly.

How can we integrate numeric approaches?

- Calculations are only approximate.
- We want an exact guarantee.

## Numeric methods

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Sicun Gao, Ed Clarke, and I proposed a framework that offers:

- More flexibility: arbitrary computable functions
- A restriction: quantification only over bounded domains
- A compromise: approximate decidability (but with an exact guarantee)

This provides a general framework for thinking about verification problems.

## Numeric methods

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Choose a language with  $0$ ,  $+$ ,  $-$ ,  $<$ ,  $\leq$ ,  $|\cdot|$ , and symbols for *any* computable functions you want.

Fix a “tolerance”  $\delta > 0$ . We defined:

- $\varphi^{+\delta}$ , a slight strengthening of  $\varphi$
- $\varphi^{-\delta}$ , a slight weakening of  $\varphi$

such that whenever  $\delta' \geq \delta \geq 0$ , we have

$$\varphi^{+\delta'} \rightarrow \varphi^{+\delta} \rightarrow \varphi \rightarrow \varphi^{-\delta} \rightarrow \varphi^{-\delta'}.$$

## Numeric methods

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Say a formula  $\varphi$  is *bounded* if every quantifier is of the form  $\forall x \in [s, t]$  or  $\exists x \in [s, t]$ .

**Theorem.** There is an algorithm which, given any bounded formula  $\varphi$ , correctly returns one of the following two answers:

- $\varphi$  is true
- $\varphi^{+\delta}$  is false.

For verification problems, think of the first answer as “the system is safe,” and the second as “a small perturbation of the system is unsafe.”

Note that there is a grey area where either answer is allowed.

## Numeric methods

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This is a theoretical result. The practical goal is to implement such an algorithm.

Gao, Clarke, Soonho Kong, and others are developing a tool, *dReal*:

- It focuses on the existential / universal fragment.
- It uses an SMT (“satisfiability modulo theories”) framework.
- It uses IBEX for interval constraint propagation.
- It uses the CAPD library to compute numerical enclosures for ODE’s.

See <https://dreal.github.io>

## A heuristic symbolic method

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Consider the following implication:

$$0 < x < y, u < v$$

$$\implies$$

$$2u + \exp(1 + x + x^4) < 2v + \exp(1 + y + y^4)$$

- This inference is not contained in linear arithmetic or real closed fields.
- This inference is tight: symbolic or numeric approximations are not useful.
- Backchaining using monotonicity properties suggests many equally plausible subgoals.
- But, the inference is completely straightforward.

## A heuristic symbolic method

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Robert Lewis and I (initially with Cody Roux) have developed a new approach, that:

- verifies inequalities on which other procedures fail
- extends beyond the language of RCF
- is amenable to producing proof terms
- captures natural patterns of inference

But:

- It is not complete.
- It not guaranteed to terminate.

It is designed to complement other procedures.

## A heuristic symbolic method

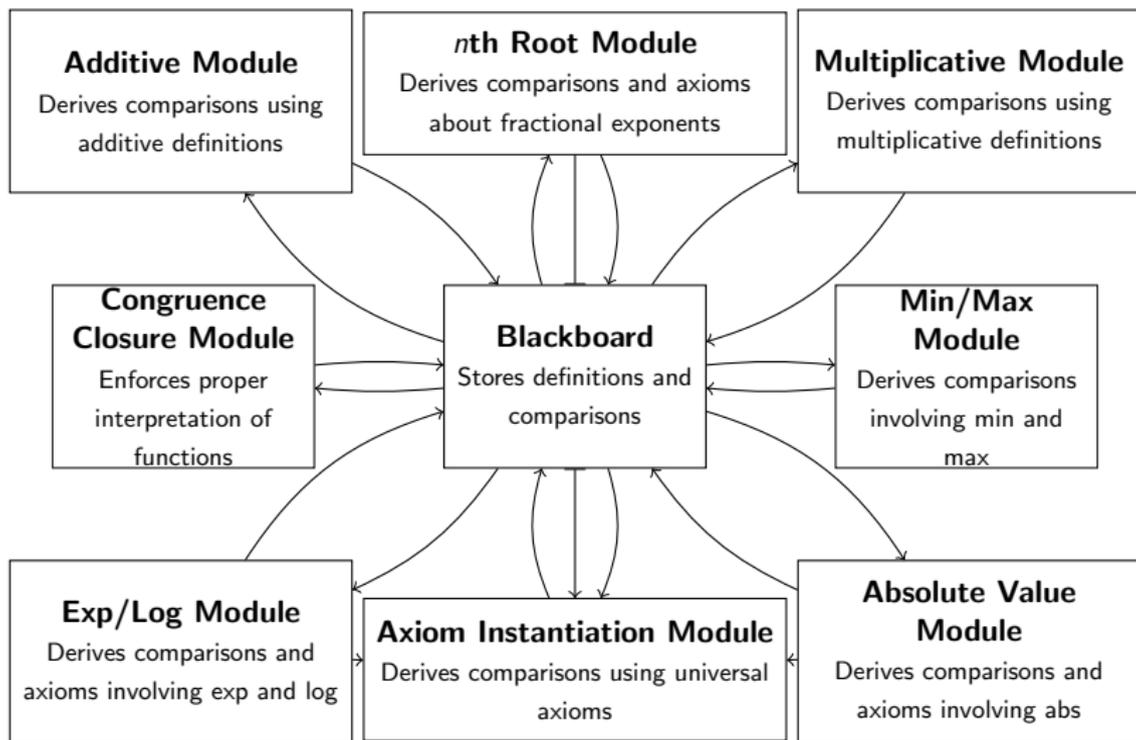
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The main ideas:

- Use forward reasoning (guided by the structure of the problem).
- Show “hypotheses  $\Rightarrow$  conclusion” by negating the conclusion and deriving a contradiction.
- As much as possible, put terms in canonical “normal forms,” e.g. to recognize that  $3(x + y)$  is a multiple of  $2y + 2x$ .
- Derive relationships between “terms of interest,” including subterms of the original problem.
- Different modules contribute bits of information, based on their expertise.

# Computational structure

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## A heuristic symbolic method

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We have a prototype Python implementation, *Polya*.

The code is open-source and available online.

- An associated paper.
- Rob's MS thesis.
- Slides from Rob's talks (from which I have borrowed).

We are planning to implement this in Lean.

# Outline

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- Formal methods
- Interactive theorem proving
- Automated reasoning
- Verified computation

## Verified computation

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Important computational proofs have been verified:

- The four color theorem (Gonthier, 2004)
- The Kepler conjecture (Hales et al., 2014)
- Various aspects of Kenzo (computation in algebraic topology) have been verified:
  - in ACL2 (simplicial sets, simplicial polynomials)
  - in Isabelle (the basic perturbation lemma)
  - in Coq/SSReflect (effective homology of bicomplexes, discrete vector fields)

# Verified computation

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Some approaches:

- rewrite the computations to construct proofs as well (Flyspeck: nonlinear bounds)
- verify certificates (e.g. proof sketches, duality in linear programming)
- verify the algorithm, and then execute it with a specialized (trusted) evaluator (Four color theorem)
- verify the algorithm, extract code, and run it (trust a compiler or interpreter) (Flyspeck: enumeration of tame graphs)

# Verified computation

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In dynamical systems:

- Immler and Hölzl have obtained verified solutions to ODE's.
- Immler has obtained verified algorithms for geometric zonotopes (for convex hull calculations).
- He has verified an enclosure for the Lorenz attractor.

The formalizations are carried out in Isabelle, and Standard ML code is extracted.

# Conclusions

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- Computers change the kinds of proofs that we can discover and verify.
- In the long run, formal methods *will* play an important role in mathematics.
- It will take clever ideas, and hard work, to understand how to use them effectively.